

Aim: testing hypothesis about mean of a normal random sample with unknown variance.

Let $X = (X_1, \dots, X_N)$ with $X_j \sim n(\mu, \sigma^2)$ ↙ ↘ unknown

be the random sample, and let $H_0: \mu = \mu_0$

for a given $\mu_0 \in \mathbb{R}$. On the other hand, $\sigma^2 > 0$

is arbitrary. Set $\Theta := (\mu, \sigma^2)$.

Recall that $L(\Theta | \underline{x}) = \left(\frac{1}{2\pi\sigma^2}\right)^{N/2} \exp\left(-\sum_{j=1}^N (x_j - \mu)^2 / 2\sigma^2\right)$.

The maximum of $\Theta \mapsto L(\Theta | \underline{x})$ is realized

for $\mu = \bar{x}$ and $\sigma^2 = \frac{1}{N} \sum_{j=1}^N (x_j - \bar{x})^2$,

see Example 7.2.11 in [CB]. By the same

argument one gets that if $\mu = \mu_0$, the maximum

of $\sigma^2 \mapsto L(\mu_0, \sigma^2 | \underline{x})$ is realized for

$\sigma^2 = \frac{1}{N} \sum_j (x_j - \mu_0)^2$. These results are obtained

by looking at the critical points of $\Theta \mapsto L(\Theta | \underline{x})$.

Then, set $\sigma_0^2 := \frac{1}{N} \sum_j (x_j - \mu_0)^2$ and $\sigma^2 := \frac{1}{N} \sum_j (x_j - \bar{x})^2$.

$$\lambda(\underline{x}) := \frac{\sup_{\sigma^2 > \sigma_0^2} L(\mu_0, \sigma^2 | \underline{x})}{\sup_{\mu, \sigma^2} L(\mu, \sigma^2 | \underline{x})}$$

$$= \frac{(2\pi\sigma_0^2)^{-N/2} \exp\left(-\sum_j (x_j - \mu_0)^2 / 2\sigma_0^2\right)}{(2\pi\sigma^2)^{-N/2} \exp\left(-\sum_j (x_j - \bar{x})^2 / 2\sigma^2\right)}$$

$$= \left(\frac{\sigma^2}{\sigma_0^2}\right)^{N/2} \frac{e^{-N/2}}{e^{-N/2}}$$

$$= \left(\frac{\sigma^2}{\sigma_0^2}\right)^{N/2}$$

$$= \left(\frac{\sum_j (x_j - \bar{x})^2}{\sum_j (x_j - \mu_0)^2}\right)^{N/2} \cdot$$

not known but fixed



Thus, one would reject H_0 if $\lambda(\underline{x}) \leq c$ for a given $c \in (0, 1)$. But let us go one more

step. Observe that $\sum_j (x_j - \mu_0)^2 = \sum_j (x_j - \bar{x})^2 + N(\bar{x} - \mu_0)^2$.

$$\text{Then } \lambda(\underline{x}) \leq c \iff \left(\frac{\sum_j (x_j - \bar{x})^2}{\sum_j (x_j - \mu_0)^2}\right)^{N/2} \leq c$$

$$\iff \frac{\sum_j (x_j - \bar{x})^2}{\sum_j (x_j - \bar{x})^2 + N(\bar{x} - \mu_0)^2} \leq c^{2/N}$$

$$\iff \frac{1}{1 + \frac{N(\bar{x} - \mu_0)^2}{\sum_j (x_j - \bar{x})^2}} \leq c^{2/N}$$

$$\Leftrightarrow \frac{N(\bar{x} - \mu_0)^2}{\sum_j (x_j - \bar{x})^2} \gg \frac{1}{c^{2/N}} - 1$$

$$\Leftrightarrow \frac{(\sqrt{N}(\bar{x} - \mu_0))^2}{\frac{1}{N-1} \sum_j (x_j - \bar{x})^2} \gg (N-1)(c^{-2/N} - 1)$$

sample mean
sample variance s^2

$$\Leftrightarrow \frac{|\bar{x} - \mu_0|}{s/\sqrt{N}} \gg \underbrace{\left((N-1)(c^{-2/N} - 1) \right)^{1/2}}_{=: c'}$$

follow a student t-distribution with parameter $N-1$.

Then, H_0 would be rejected if $\frac{|\bar{x} - \mu_0|}{s/\sqrt{N}} \gg c'$.

Conclusion: The likelihood ratio test is equivalent to a t-test (in this setting).

Some useful exercises:

Example 7.2.11 + 7.2.12

Example 8.2.3

Thm 8.2.4 about sufficient statistics

Exercise 8.3

Example 8.2.9